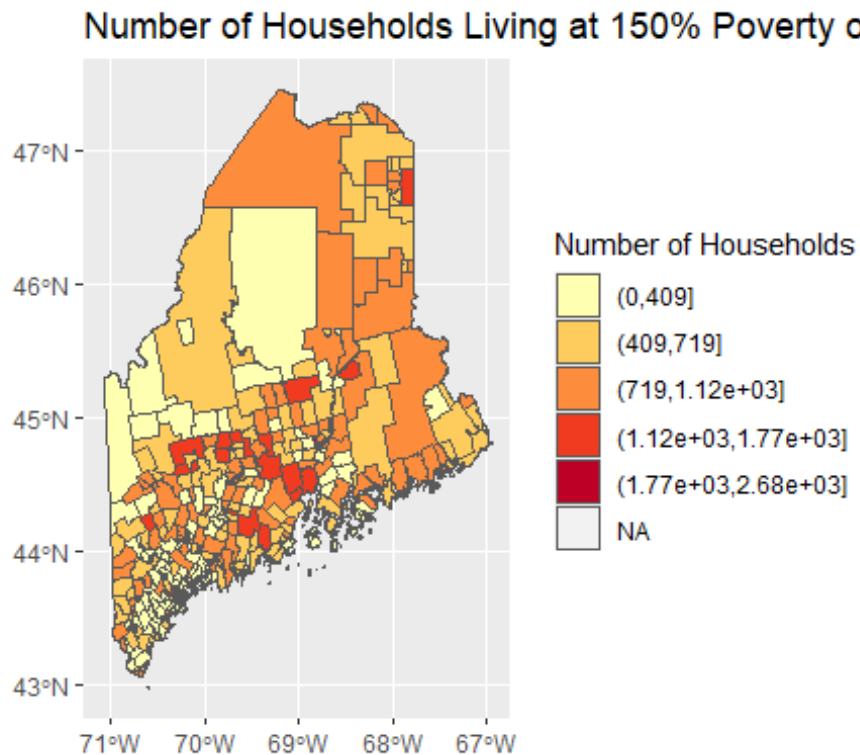


```
brks=classIntervals(mainesv$E_POV150, n=5, style="jenks", intervalClosure = '
left')
mainesv$pov <- cut(mainesv$E_POV150, brks$brks)

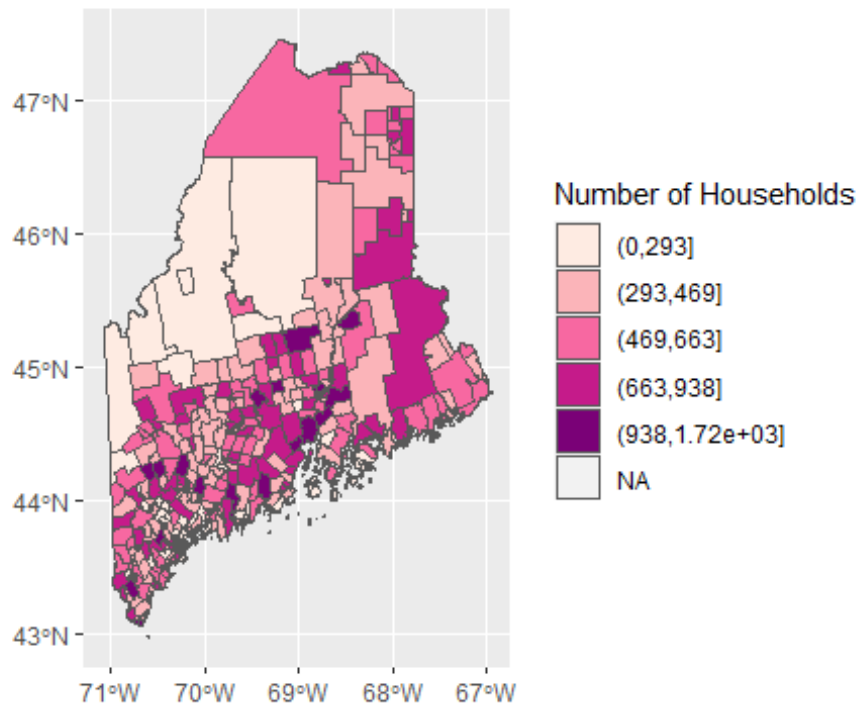
ggplot() +
  geom_sf(data = mainesv, aes(fill = pov)) +
  scale_fill_brewer(palette = "YlOrRd") +
  coord_sf() +labs(title="Number of Households Living at 150% Poverty or Les
s", size=1, fill="Number of Households")
```



```
brks=classIntervals(mainesv$E_DISABL, n=5, style="jenks", intervalClosure = '
left')
mainesv$nohs <- cut(mainesv$E_DISABL, brks$brks)

ggplot() +
  geom_sf(data = mainesv, aes(fill = nohs)) +
  scale_fill_brewer(palette = "RdPu") +
  coord_sf() +labs(title="Number of Individuals with Documented Disability",
size=1, fill="Number of Households")
```

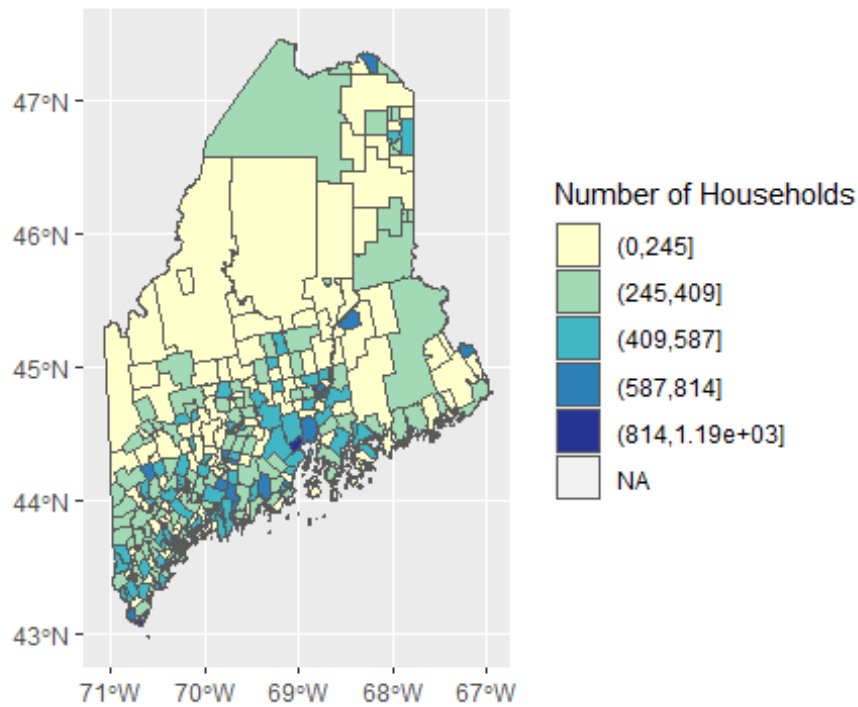
Number of Individuals with Documented Disability



```
brks=classIntervals(mainesv$E_HBURD, n=5, style="jenks", intervalClosure = 'left')
mainesv$hburd <- cut(mainesv$E_HBURD, brks$brks)

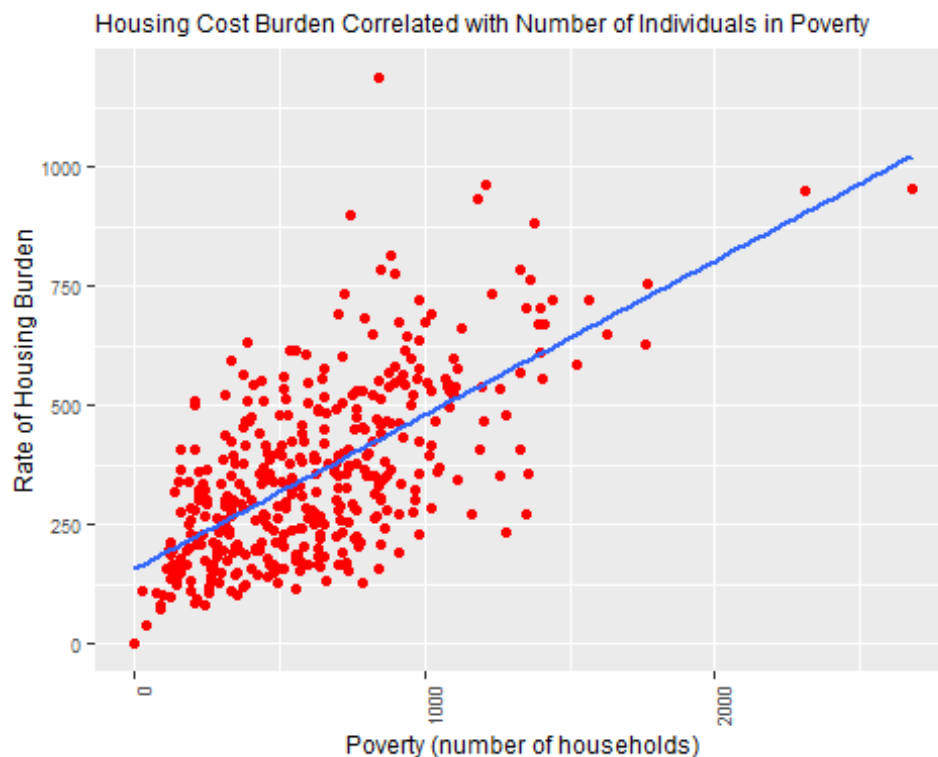
ggplot() +
  geom_sf(data = mainesv, aes(fill = hburd)) +
  scale_fill_brewer(palette = "YlGnBu") +
  coord_sf() +labs(title="Number of Households With Housing Cost Burden", size=1, fill="Number of Households")
```

Number of Households With Housing Cost Burden



```
ggplot(data = mainesv, aes(x = E_POV150, y=E_HBURD)) + geom_point(col='red')+
geom_smooth(method=lm, se=FALSE)+ labs(title = "Housing Cost Burden Correlated
with Number of Individuals in Poverty", x = "Poverty (number of households)",
y = "Rate of Housing Burden")+ theme(plot.title = element_text(size=9), axis
.text.x=element_text(angle=90, hjust=1, size = 7), axis.title.x =element_text
(size =9), axis.text.y=element_text(hjust=1, size = 7), axis.title.y =element
_text(size =9))
```

```
## `geom_smooth()` using formula 'y ~ x'
```



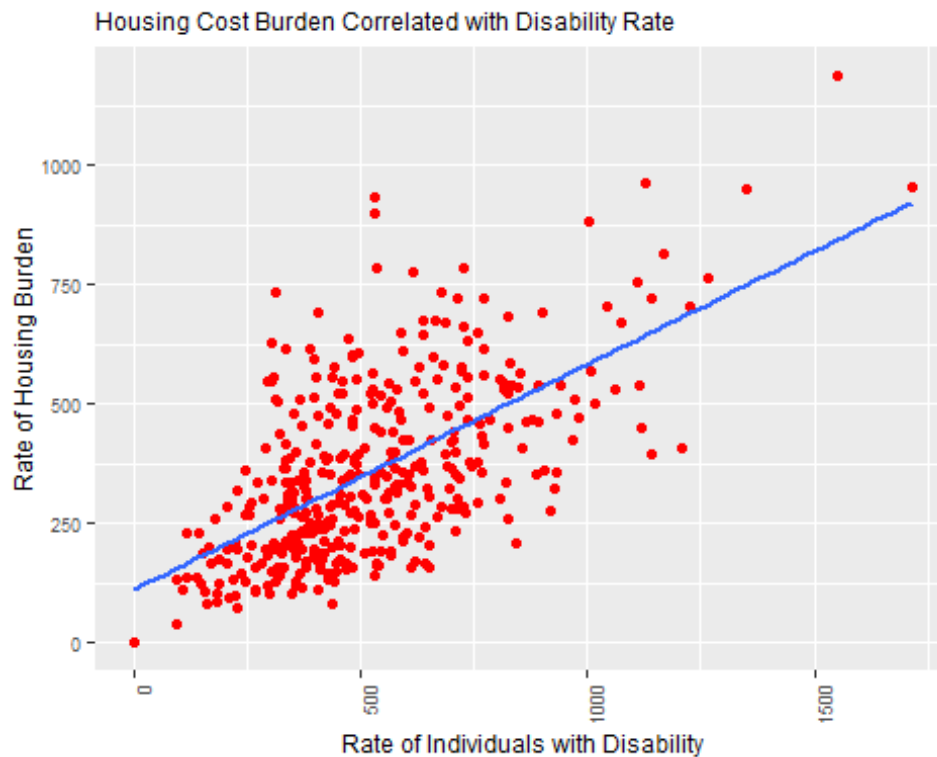
```
lmpov <- lm(E_HBURD ~ E_POV150, data=mainesv)
summary(lmpov)
```

```
##
## Call:
## lm(formula = E_HBURD ~ E_POV150, data = mainesv)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -334.69 -101.79  -16.68   86.20  758.97
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 155.69773   14.16576   10.99  <2e-16 ***
## E_POV150     0.32344    0.01956   16.53  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 142.8 on 399 degrees of freedom
## Multiple R-squared:  0.4066, Adjusted R-squared:  0.4051
## F-statistic: 273.4 on 1 and 399 DF, p-value: < 2.2e-16
```

```
ggplot(data = mainesv, aes(x = E_DISABL, y=E_HBURD)) + geom_point(col='red')+
geom_smooth(method=lm, se=FALSE)+ labs(title="Housing Cost Burden Correlated
with Disability Rate", x = "Rate of Individuals with Disability", y = "Rate o
f Housing Burden")+ theme(plot.title = element_text(size=9), axis.text.x=ele
ment_text(angle=90, hjust=1, size = 7), axis.title.x =element_text(size =9),
```

```
axis.text.y=element_text(hjust=1, size = 7), axis.title.y =element_text(size
=9))
```

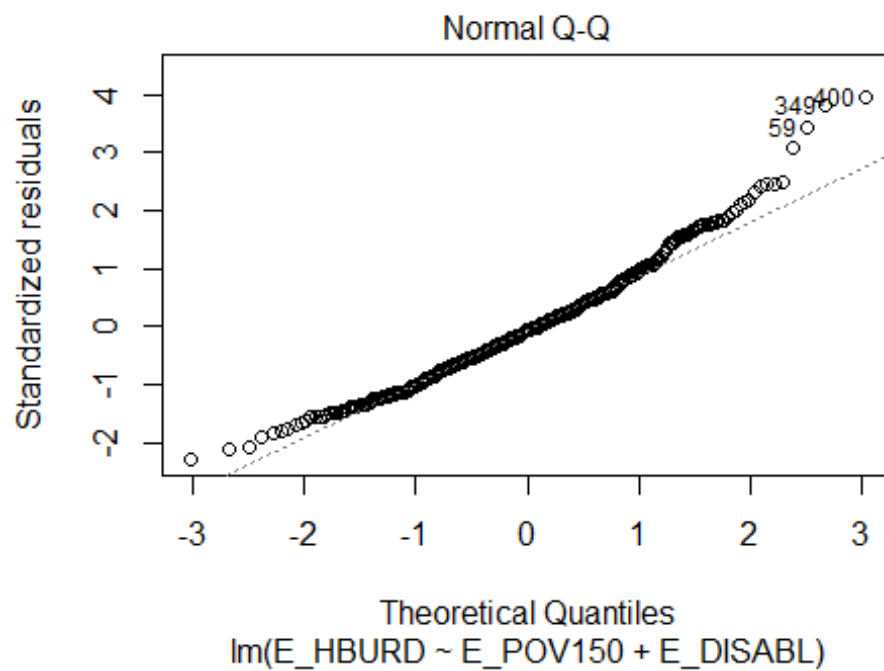
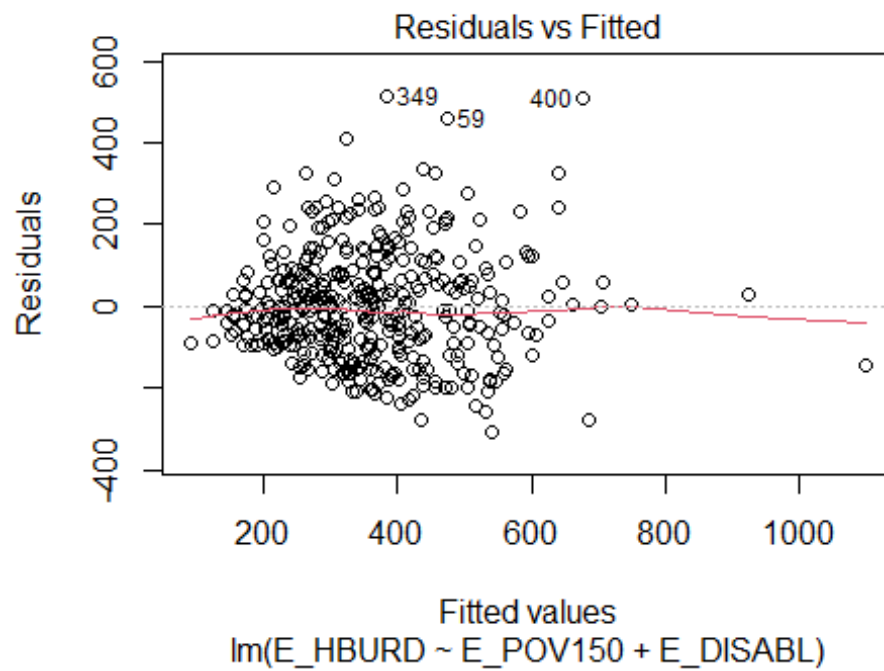
```
## `geom_smooth()` using formula 'y ~ x'
```

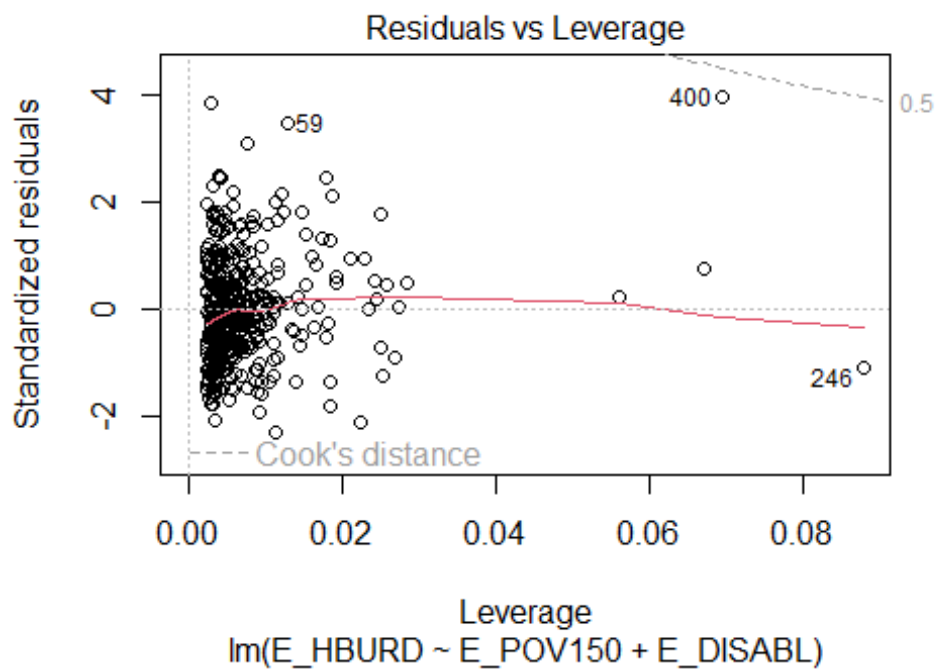
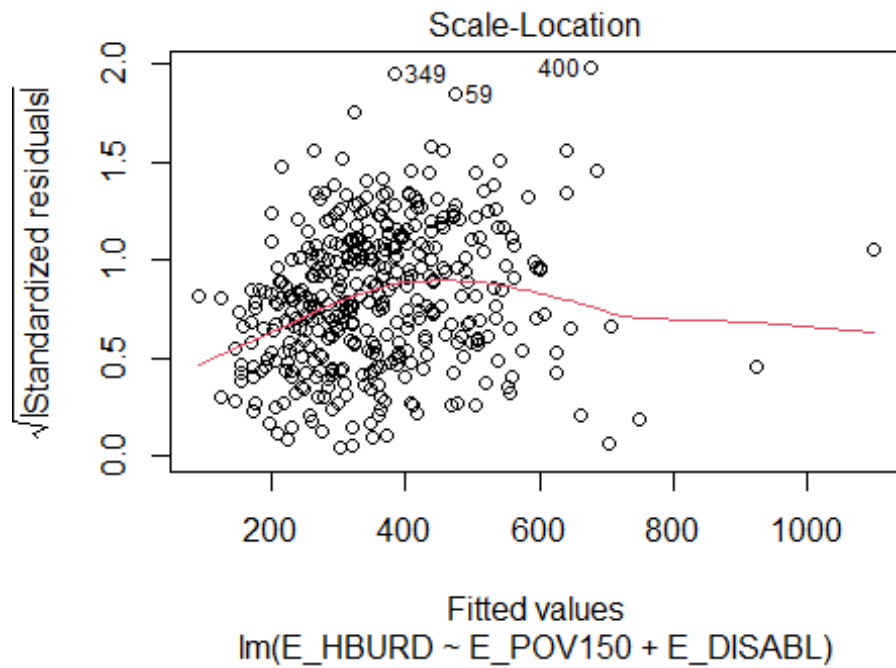


```
lmnohs <- lm(E_HBURD ~ E_DISABL, data=mainesv)
summary(lmnohs)
```

```
##
## Call:
## lm(formula = E_HBURD ~ E_DISABL, data = mainesv)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -301.34 -102.44  -28.51   76.29  573.96
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 110.66502   17.32856   6.386 4.73e-10 ***
## E_DISABL      0.47230    0.03002  15.730 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 145.6 on 399 degrees of freedom
## Multiple R-squared:  0.3828, Adjusted R-squared:  0.3812
## F-statistic: 247.4 on 1 and 399 DF, p-value: < 2.2e-16
```

```
ols <- lm(E_HBURD ~ E_POV150+E_DISABL, data=mainesv)
plot(ols)
```





```
coef(ols)
```

```
## (Intercept)    E_POV150    E_DISABL
##  90.0448517    0.2056088    0.2660604
```

```

summary(ols)

##
## Call:
## lm(formula = E_HBURD ~ E_POV150 + E_DISABL, data = mainesv)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -306.59  -91.91   -9.34   76.63  513.76
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  90.04485    16.20638    5.556 5.06e-08 ***
## E_POV150      0.20561     0.02473    8.313 1.49e-15 ***
## E_DISABL      0.26606     0.03722    7.148 4.24e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 134.6 on 398 degrees of freedom
## Multiple R-squared:  0.4741, Adjusted R-squared:  0.4715
## F-statistic: 179.4 on 2 and 398 DF,  p-value: < 2.2e-16

bptest(ols)

##
## studentized Breusch-Pagan test
##
## data:  ols
## BP = 16.872, df = 2, p-value = 0.0002169

jarque.bera.test(ols$residuals)

##
## Jarque Bera Test
##
## data:  ols$residuals
## X-squared = 36.587, df = 2, p-value = 1.135e-08

nbscali=poly2nb(mainesv, queen = TRUE)
W <- nb2listw(nbscali, style = "W", zero.policy=TRUE)

lm.morantest(ols, W, zero.policy=TRUE)

##
## Global Moran I for regression residuals
##
## data:
## model: lm(formula = E_HBURD ~ E_POV150 + E_DISABL, data = mainesv)
## weights: W
##
## Moran I statistic standard deviate = 13.078, p-value < 2.2e-16

```



```

## alternative hypothesis: greater
## sample estimates:
## Observed Moran I      Expectation      Variance
##      0.4052065868      -0.0032494078      0.0009754505

moran.test(ols$residuals, W, randomisation = FALSE, zero.policy=TRUE)

##
## Moran I test under normality
##
## data:  ols$residuals
## weights: W  n reduced by no-neighbour observations
##
## Moran I statistic standard deviate = 13.028, p-value < 2.2e-16
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic      Expectation      Variance
##      0.4052065868      -0.0025125628      0.0009794059

moran.test(ols$residuals, W, randomisation = TRUE, zero.policy=TRUE)

##
## Moran I test under randomisation
##
## data:  ols$residuals
## weights: W  n reduced by no-neighbour observations
##
## Moran I statistic standard deviate = 13.04, p-value < 2.2e-16
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic      Expectation      Variance
##      0.4052065868      -0.0025125628      0.0009775584

moran.mc(ols$residuals, W, 99, zero.policy=TRUE)

##
## Monte-Carlo simulation of Moran I
##
## data:  ols$residuals
## weights: W
## number of simulations + 1: 100
##
## statistic = 0.40521, observed rank = 100, p-value = 0.01
## alternative hypothesis: greater

moran.test(ols$residuals, W, randomisation = FALSE, zero.policy = TRUE)

##
## Moran I test under normality
##

```

```

## data:  ols$residuals
## weights: W  n reduced by no-neighbour observations
##
##
## Moran I statistic standard deviate = 13.028, p-value < 2.2e-16
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic      Expectation      Variance
##      0.4052065868      -0.0025125628      0.0009794059

logLik(ols)

## 'log Lik.' -2533.202 (df=4)

BIC(ols)

## [1] 5090.38

lmtests=lm.LMtests(ols, W, zero.policy = T, test='all')
lstat=unlist(c(lmtests$LMerr[1],lmtests$LMlag[1], lmtests$RLMerr[1],lmtests$R
LMlag[1], lmtests$SARMA[1]))
pval=(c(lmtests$LMerr[3],lmtests$LMlag[3], lmtests$RLMerr[3],lmtests$RLMlag[3
], lmtests$SARMA[3]))

summary(lmtests)

## Lagrange multiplier diagnostics for spatial dependence
## data:
## model: lm(formula = E_HBURD ~ E_POV150 + E_DISABL, data = mainesv)
## weights: W
##
##      statistic parameter    p.value
## LMerr      166.997          1 < 2.2e-16 ***
## LMlag       62.357          1 2.887e-15 ***
## RLMerr     116.640          1 < 2.2e-16 ***
## RLMlag      12.001          1 0.0005317 ***
## SARMA      178.998          2 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

serror<- errorsarlm(E_HBURD ~ E_POV150 + E_DISABL, data = mainesv, listw=W, z
ero.policy = T)
summary(serror)

##
## Call:errorsarlm(formula = E_HBURD ~ E_POV150 + E_DISABL, data = mainesv,
##      listw = W, zero.policy = T)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -345.139  -67.405  -12.635   56.047  443.534
##

```

```

## Type: error
## Regions with no neighbours included:
## 162 343
## Coefficients: (asymptotic standard errors)
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept) 65.113054 17.937467 3.6300 0.0002834
## E_POV150     0.237907  0.021707 10.9599 < 2.2e-16
## E_DISABL     0.279476  0.031521  8.8663 < 2.2e-16
##
## Lambda: 0.56285, LR test value: 108.78, p-value: < 2.22e-16
## Asymptotic standard error: 0.053446
##      z-value: 10.531, p-value: < 2.22e-16
## Wald statistic: 110.91, p-value: < 2.22e-16
##
## Log likelihood: -2478.812 for error model
## ML residual variance (sigma squared): 12766, (sigma: 112.99)
## Number of observations: 401
## Number of parameters estimated: 5
## AIC: 4967.6, (AIC for lm: 5074.4)

slag<- lagsarlm(E_HBURD ~ E_POV150 + E_DISABL, data = mainesv, listw=W, zero.
policy = T)
summary(slag)

##
## Call:lagsarlm(formula = E_HBURD ~ E_POV150 + E_DISABL, data = mainesv,
##      listw = W, zero.policy = T)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -300.152  -85.296  -15.787   71.345  503.094
##
## Type: lag
## Regions with no neighbours included:
## 162 343
## Coefficients: (asymptotic standard errors)
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept) -38.500599 20.416091 -1.8858 0.05932
## E_POV150     0.196379  0.023175  8.4736 < 2.2e-16
## E_DISABL     0.266364  0.034358  7.7527 9.104e-15
##
## Rho: 0.37376, LR test value: 51.403, p-value: 7.524e-13
## Asymptotic standard error: 0.048814
##      z-value: 7.6567, p-value: 1.9096e-14
## Wald statistic: 58.625, p-value: 1.9096e-14
##
## Log likelihood: -2507.501 for lag model
## ML residual variance (sigma squared): 15367, (sigma: 123.96)
## Number of observations: 401
## Number of parameters estimated: 5

```

```
## AIC: 5025, (AIC for lm: 5074.4)
## LM test for residual autocorrelation
## test value: 43.172, p-value: 5.0146e-11

p <- c(AIC(ols), AIC(slag), AIC(serror))
q <- c(BIC(ols), BIC(slag), BIC(serror))
r <- c(logLik(ols), logLik(slag), logLik(serror))

labdata <- data.frame(p,q,r)
names(labdata) <- c("AIC", "BIC", "Log Likelihood")
rownames(labdata) <-c("OLS Model", "Spatial Lag Model", "Spatial Error Model"
)
print(labdata)

##              AIC      BIC Log Likelihood
## OLS Model      5074.404 5090.380      -2533.202
## Spatial Lag Model  5025.002 5044.972      -2507.501
## Spatial Error Model 4967.624 4987.594      -2478.812

summary(labdata)

##      AIC      BIC      Log Likelihood
## Min.   :4968  Min.   :4988  Min.      :-2533
## 1st Qu.:4996  1st Qu.:5016  1st Qu.:-2520
## Median :5025  Median :5045  Median :-2508
## Mean   :5022  Mean   :5041  Mean    :-2507
## 3rd Qu.:5050  3rd Qu.:5068  3rd Qu.:-2493
## Max.   :5074  Max.   :5090  Max.     :-2479
```